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A VARIATIONAL EQUIVALENT TO DIAGONAL SCALING

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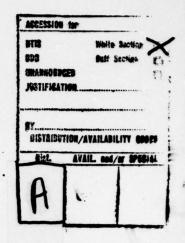
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ABSTRACT



This paper is concerned with the problem of diagonally scaling a given nonnegative matrix a to one with prescribed row and column sums. The approach is to represent one of the two scaling matrices as the solution of a variational problem. This leads in a natural way to necessary and sufficient conditions on the zero pattern of a so that such a scaling exists. In addition the convergence of the successive prescribed row and column sum normalizations is established. Certain invariants under diagonal scaling are used to actually compute the desired scaled matrix, and examples are provided. Finally, at the end of the paper, a discussion of infinite systems is presented.

AMS(MOS) Subject Classification: 15A21, 26A54.

Key Words: Diagonal scaling, Jensen's inequality, Weak convergence.

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SIGNIFICANCE AND EXPLANATION

In order to avoid heavy round-off error when solving large systems of linear equations on the computer, it is useful to scale the coefficient matrix. This scaling amounts to multiplying the matrix by diagonal matrices on the left and right. We consider nonnegative coefficients and concern ourselves with the problem of choosing appropriate diagonal matrices in order that the scaled matrix have certain prescribed row and column sums. In addition, if some of the coefficients are zero, we ask if this can at all be done.

Our approach is to replace the scaling problem by an equivalent problem of finding the minimum of a certain function. This facilitates much of the analysis and leads in a natural way to necessary and sufficient conditions on the pattern of zeros in the coefficient matrix. Included is a technique for actually computing the appropriate scaling matrices, a number of examples illustrating the results, and a discussion of infinite systems.

The responsibility for the wording and views expressed in this descriptive summary lies with MRC, and not with the authors of this report.

A VARIATIONAL EQUIVALENT TO DIAGONAL SCALING

Marc A. Berger and C. T. Kelley

Introduction

In this paper we present a new approach to the problem of diagonally scaling a non-negative man matrix a to one with prescribed row and column sums σ and δ ; and to the question of convergence of the successive σ and δ row and column sum normalizations of a. Our methods show this equivalent to the problem of minimizing

$$\prod_{j=1}^{n} \left(\sum_{i=1}^{m} a_{ij} t_{i} \right)^{j}$$

over R (positive vectors in R) subject to

$$\prod_{i=1}^{m} t_i^{\sigma_i} = 1.$$

And this leads us in a natural way to conditions on the pattern of zero elements in a (Theorem 2.D). The main computational result underlying the theoretical parts of the paper states that the solution to the above minimum problem is given by (3.12).

In §1 we present the problem and the tools employed to solve it. In §2 the existence, non-existence, uniqueness and convergence results are presented for finite systems. In §3 certain invariants under diagonal scaling are described, and used to actually compute the desired scaled matrix. And some examples are provided. Finally, in the last section a discussion of infinite systems, using functional analysis, is presented.

The results in §2 concerning the scaling and convergence questions are not new. The original work involved doubly stochastic matrices, and appears in Sinkhorn [7], Sinkhorn and Knopp [8], Menon [5], and Brualdi, Parter and Schneider [3]. The generalization to non-square matrices and arbitrary positive σ and δ appears in Bacharach [1], Brualdi [2], and Menon and Schneider [6].

The approach of representing the solution in terms of a variational problem is used by Theil [9] and Gorman [4]. The former minimizes

over positive matrices $\,b\,$ having $\,\sigma\,$ and $\,\delta\,$ as row and column sums. And the latter minimizes

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$$\sum_{i=1}^{m}\sum_{j=1}^{n} a_{ij} e^{t_i+s_j} - \sum_{i=1}^{m} \sigma_i t_i - \sum_{j=1}^{m} \delta_j s_j$$

over R^{m+n}₊. The problem we formulate, however, is tied directly to the Menon operator in [5], and leads in a most natural and direct way to the full admissibility condition of Theorem 2.D.

§1. The Scaling Problem

Let σ and δ be positive vectors in \mathbb{R}^m and \mathbb{R}^n , respectively, satisfying

And let $\mathcal Q$ denote the set of nonnegative matrices a in $\mathbb R^m \otimes \mathbb R^n$ with no vanishing rows or columns. Define normalizing operators $\mathbb R$ and $\mathbb C$ from $\mathcal Q$ into itself by

(1.2)
$$(Ra)_{ij} = a_{ij} \left(\sum_{k=1}^{n} a_{ik} \right)^{-1} \sigma_{i}, (Ca)_{ij} = a_{ij} \left(\sum_{k=1}^{m} a_{kj} \right)^{-1} \delta_{j}.$$

Let

$$(1.3) \quad \Lambda = CR$$

The reader can verify that

$$\mathfrak{F}(\Lambda) = \mathfrak{F}(C) \cap \mathfrak{F}(R) \equiv \mathfrak{F}(C,R)$$

over the set $\mathbb{R}^{m,n}_+$ of positive matrices in $\mathbb{R}^m \otimes \mathbb{R}^n$, where \mathfrak{F} denotes the set of fixed points. Of course, over \mathcal{Q} , $\mathfrak{F}(\Lambda)$ is strictly larger than $\mathfrak{F}(C,\mathbb{R})$. We concern ourselves with the following two problems.

For which a ϵQ does the sequence

- $\{\Lambda^{N} a\}$ converge to a limit in $\mathfrak{F}(C,R)$, and to what limit?
- For which a $\in \mathcal{Q}$ do there exist positive diagonal matrices \mathbf{x} and \mathbf{y} (P₂) such that $\mathbf{xay} \in \mathcal{F}(C,R)$, and are they unique up to scalar multiples?

Let $\mathbb{R}^{\mathfrak{m}}_{+}$ denote the set of positive vectors t in $\mathbb{R}^{\mathfrak{m}}$. For a fixed a $\in \mathcal{Q}$ we associate an operator $\lambda \equiv \lambda(a)$ from $\mathbb{R}^{\mathfrak{m}}_{+}$ into itself by

(1.5)
$$(\lambda t)_{i} = \left[\sum_{j=1}^{n} a_{ij} \delta_{j} \left(\sum_{k=1}^{m} a_{kj} t_{k} \right)^{-1} \right]^{-1} \sigma_{i}$$

This operator was defined by Menon in [5], and was used by Brualdi, Parter and Schneider in [3] to analyze (P_2) when $\mathcal{F}(C,R)$ is the assignment polytope Ω_n . It was also used by Menon and Schneider in [6] to analyze (P_2) in the more general setting put forth here.

We note that if $\lambda t=t$ then the matrix b defined by

(1.6)
$$b_{ij} = a_{ij} \left(\sum_{k=1}^{m} a_{kj} t_{k} \right)^{-1} t_{i} \delta_{j}$$

is in $\Im(C,R)$. And conversely, if b of the form $b_{ij} = a_{ij}t_{ij}$ is in $\Im(C,R)$, then $\lambda t = t$. Furthermore,

(1.7)
$$(\Lambda^{N}a)_{ij} = a_{ij} \left[\sum_{k=1}^{m} a_{kj} (\lambda^{N-1}t^{0})_{k} \right]^{-1} (\lambda^{N-1}t^{0})_{i} \delta_{j}$$

where

(1.8)
$$t_{i}^{0} = \left(\sum_{j=1}^{n} a_{ij} \right)^{-1} \sigma_{i}.$$

And by inverting this it follows that

(1.9)
$$(\lambda^{N} t^{0})_{i} = \sigma_{i}^{N+1} \prod_{k=0}^{N} [\sum_{j=1}^{n} (\lambda^{k} a)_{ij}]^{-1}$$

Thus λ allows one to consider both (P_1) and (P_2) .

We show in the next section that solving \alphat=t corresponds to minimizing

n m δ_j m $-\sigma_j$ T t δ_j m $-\sigma_j$ T t δ_j T t t δ_j T t δ_j T t t δ_j

§2. Existence - Uniqueness - Convergence

Consider the function Φ defined on \mathbb{R}^{m}_{+} by

(2.1)
$$\phi(t) = \prod_{j=1}^{n} \left(\sum_{i=1}^{m} a_{ij} t_{i} \right)^{\delta_{j}} \prod_{i=1}^{m} t_{i}^{-\sigma_{i}} .$$

The condition $\nabla \log \phi(t) = 0$ is equivalent to $\lambda t = t$. Thus if ϕ has a critical point in \mathbb{R}^m_+ then, in accord with the remarks in the previous section, there exist positive diagonal matrices x and y such that $xay \in \mathcal{F}(C,R)$. We consider the problem of minimizing ϕ . It follows from (1.1) that ϕ is homogeneous in t, and thus it suffices to impose the constraint $t \in B$, where

(2.2)
$$B = \{t \in \mathbb{R}^{m}_{+} : \sum_{i=1}^{m} t_{i} = 1\}$$

If $a \in \mathbb{R}_+^{m,n}$ then ϕ tends to infinity as t approaches the boundary ∂B from inside of B. And since ϕ is continuous in B, it follows that a minimum exists in B. However, if some of the elements of a vanish, then this need not be the case. In fact, if F is a proper subset of $\mathbf{Z}_m = \{1, \dots, m\}$ and F' is its complement, we can examine the behavior of $\phi(t^* + \epsilon \tau_{\epsilon})$ as ϵ tends to zero; where t_i^* is zero for $i \in F$ and positive for $i \in F'$, and $\tau_{\epsilon} \in \mathbb{R}_+^m$ with $\tau_{\epsilon,i} = 1$ for $i \in F$. The condition that the limit be infinite is

(2.3)
$$\sum_{\mathbf{j} \in C(\mathbf{F})} \delta_{\mathbf{j}} < \sum_{\mathbf{i} \in \mathbf{F}} \sigma_{\mathbf{i}}$$

where

(2.4)
$$C(F) = \{ j : a_{kj} = 0 \text{ for each } k \in F' \}$$
.

Furthermore, note that C(F) is empty unless F contains one of the sets

(2.5)
$$F_{\ell} = \{i : a_{i\ell} > 0\}$$

Theorem 2.A:

Let a $\in \mathcal{A}$ be such that (2.3) holds for all proper subsets F which are unions of the sets F_{ℓ} in (2.5). Then there exist positive diagonal matrices x and y such that xay $\in \mathfrak{F}(C,R)$, and they are unique up to scalar multiples.

Proof:

First we establish (2.3) for any proper subset F. Let F be the union of the F_{ℓ} contained in F. Then F \subset F and C(F) = C(F). Therefore (2.3) must hold for F, since it holds for F .

Let $t^* \in \partial B$. Then $\lim_{\epsilon \downarrow 0} \Phi(t^* + \epsilon \tau_{\epsilon}) = \infty$ whenever $t^* + \tau_{0} \in \mathbb{R}^m_+$. Thus it follows that $\lim_{\epsilon \downarrow 0} \Phi(t) = \infty$. This means that Φ has a minimum in B , and thus that x and y t t texist.

To establish uniqueness it suffices to show that if \tilde{a} and $\tilde{x}\tilde{a}\tilde{y}$ are in $\mathfrak{F}(C,R)$ where \tilde{x} and \tilde{y} are positive diagonal matrices then $\tilde{a} = \tilde{x}\tilde{a}\tilde{y}$.

Scale \hat{x} and \hat{y} so that each $\hat{x}_i \leq 1$ and each $\hat{y}_j \geq 1$, and at least one element is equal to one. Say $\hat{x}_i = 1$. Then it follows from (2.6) that $\hat{y}_j = 1$ whenever $\hat{a}_{i_0 j} > 0$. Thus $\hat{a}_{i_0 j} = \hat{a}_{i_0 j} \hat{x}_i \hat{y}_j$ for every j, and moreover, some $\hat{y}_j = 1$. Then similarly $\hat{a}_{i_0 j} = \hat{a}_{i_0 j} \hat{x}_i \hat{y}_j$ for every j. That is, the j_0 -th row and j_0 -th column of $\hat{a}_i - \hat{x}_i \hat{a}_j \hat{y}_i$ vanish. So we delete them and consider the (m-1) \times (n-1) case. Continuing this reduction, eventually either one or both of \hat{x}_i and \hat{y}_i reduce to a single element. And this case is trivial.

Theorem 2.B:

Under the conditions of Theorem 2.A, the sequence $\{\Lambda^N a\}$ converges to xay. Proof:

It follows from (1.7) and the uniqueness result that the convergence of $\{\lambda^N t^0\}$ to an interior fixed point t of λ , implies the convergence of $\{\Lambda^N a\}$ to xay. Suppose first that there exists p such that $(aa)^p \in \mathbb{R}^m_+$, where a denotes the transpose of a. Then for any distinct vectors u and v in \mathbb{R}^m_+ with $u \leq v$, we not only have $\lambda u \leq \lambda v$, but $(\lambda^p u)_i < (\lambda^p v)_i$ for every i. Now let t be an interior fixed point of t, and let $[\alpha_0, \beta_0]$ be the smallest interval such that $\alpha_0 t \leq t^0 \leq \beta_0 t$. Then $\alpha_0 t \leq \lambda t^0 \leq \beta_0 t$ and we can reduce the interval $[\alpha_0, \beta_0]$ to the smallest $[\alpha_1, \beta_1]$ such that $\alpha_1 t \leq \lambda t^0 \leq \beta_1 t$. Continuing in this manner we generate a sequence of nested intervals $[\alpha_N, \beta_N]$ such that

$$\alpha_{N}t \leq \lambda^{N}t^{0} \leq \beta_{N}t$$

and $[\alpha_N^{},\beta_N^{}]$ is the smallest such interval. This nested sequence converges to $[\alpha,\beta]$, and it suffices now to show that $\alpha=\beta$.

So suppose $\alpha \neq \beta$ and let $t^* = \lim_{k \to \infty} {}^{N}_{k} t^0$ be any cluster point of the sequence $\{\lambda^N t^0\}$. Then $\alpha t \leq t^* \leq \beta t$ and $[\alpha, \beta]$ is the smallest such interval. But then

 $\hat{t} = \lim_{k \to \infty} {}^{N}_{k} {}^{+P}_{t} 0$ is another cluster point, and $\alpha t_{i} < \hat{t}_{i} < \beta t_{i}$ for every i; and this contradicts the minimality of $[\alpha,\beta]$. Thus $\alpha = \beta$.

Suppose next that no powers of $a\bar{a}$ are positive. Then, since $a\bar{a}$ has a positive diagonal, it can be written as a direct sum of blocks $b_{\bar{a}}$ over subspaces

$$\mathbb{R}_{\mathbf{q}}^{\mathbf{m}} = \{ \mathbf{t} \in \mathbb{R}^{\mathbf{m}} : \mathbf{t}_{\mathbf{i}} = 0 , \mathbf{i} \in \mathbb{F}_{\mathbf{q}} \}$$

where $\mathbf{F}_{\mathbf{q}} \subset \mathbf{Z}_{\mathbf{m}}$, and each $\mathbf{b}_{\mathbf{q}}$ has a positive power. And λ can accordingly be decomposed into operators $\lambda_{\mathbf{q}}$ acting on $\mathbf{R}_{\mathbf{q},+}^{\mathbf{m}}$, the set of positive vectors in $\mathbf{R}_{\mathbf{q}}^{\mathbf{m}}$. Thus it suffices to establish the convergence of $\{\lambda_{\mathbf{q}}^{\mathbf{N}} t^{0,\mathbf{q}}\}$, where $t^{0,\mathbf{q}}$ is the corresponding restriction of t^{0} to $\mathbf{R}_{\mathbf{q}}^{\mathbf{m}}$. Furthermore, this will follow immediately from the first part of this proof, as long as each $\lambda_{\mathbf{q}}$ has fixed points in $\mathbf{R}_{\mathbf{q},+}^{\mathbf{m}}$. And this is immediate from Theorem 2.A, which establishes fixed points for λ in $\mathbf{R}_{\mathbf{q},+}^{\mathbf{m}}$.

Theorems 2.A and 2.B solve (P_1) and (P_2) under conditions (2.3). In particular we note that under these conditions there always exists a matrix $b \in \mathcal{F}(C,R)$ such that $a_{ij} = 0$ if and only if $b_{ij} = 0$. What we show below is that if there exists any F such that

(2.8)
$$\sum_{j \in C(F)} \delta_j > \sum_{i \in F} \sigma_i$$

then no such b exists, even if we allow b to have extra zeros. And therefore for these a $\in \mathcal{Q}$ no diagonal matrices x and y exist for (P_2) . And $\{\Lambda^N a\}$ does not converge to a limit in $\mathcal{F}(C,R)$.

Theorem 2.C:

Suppose $a \in \mathcal{Q}$ is such that (2.8) holds for some F. Then for no $b \in \mathcal{Q} \cap \mathfrak{F}(C,R)$ is it true that $a_{ij} = 0$ implies $b_{ij} = 0$.

Proof:

Suppose such a b exists. Then (2.8) holds for b even if b has more zeros than a , because this will only serve to enlarge C(F). But then

$$\sum_{\mathbf{i} \in F} \sigma_{\mathbf{i}} = \sum_{\mathbf{i} \in F} \sum_{\mathbf{j}=1}^{m} b_{\mathbf{i} \cdot \mathbf{j}} \ge \sum_{\mathbf{i} \in F} \sum_{\mathbf{j} \in C(F)} b_{\mathbf{i} \cdot \mathbf{j}} = \sum_{\mathbf{j} \in C(F)} \sum_{\mathbf{i} \in F} b_{\mathbf{i} \cdot \mathbf{j}} = \sum_{\mathbf{j} \in C(F)} \sum_{\mathbf{i} = 1}^{m} b_{\mathbf{i} \cdot \mathbf{j}} = \sum_{\mathbf{j} \in C(F)} \delta_{\mathbf{j}}$$

which is a contradiction.

Finally we wish to consider the case where

(2.9)
$$\sum_{j \in C(F)}^{\delta_j} \delta_j = \sum_{i \in F}^{\sigma_i} \sigma_i$$

for some proper subset(s) F, (2.3) holding for the others. In this case in order that x and y solving (P_2) exist, it is necessary that

(2.10)
$$a_{ij} = 0 , i \in F , j \in C(F)'$$

for those F satisfying (2.9). This is simply because it is necessary that $x_i a_{ij} y_j = 0$,

i ϵ F , j ϵ C(F)'. The result below shows that these conditions are sufficient as well. In order to avoid cumbersome hypotheses we summarize these conditions in a definition. Thus a ϵ Q will be called <u>fully admissible</u> if (2.3) or (2.9) hold for every proper subset F; and (2.10) holds for those F satisfying (2.9).

a $\in \mathcal{Q}$ is fully admissible if and only if there exist diagonal matrices \times and Y solving (P₂). And under these conditions $\{\Lambda^N_a\}$ converges to \times ay.

Theorem 2.D:

Sufficiency has already been established in the discussion above. To prove necessity we can assume (2.9) holds for some proper subset F. Otherwise the existence of x and y follows directly from Theorem 2.A. Because of (2.10) we can reduce the problem to the two sub-matrices $(a_{ij}:i\in F,j\in C(F))$ and $(a_{ij}:i\in F',j\in C(F)')$. Both of these sub-matrices are fully admissible and their dimensions are strictly less than those of a . Thus an induction establishes the existence of diagonal matrices x' and y', x" and y" solving (P_2) for these smaller matrices. And x',x" and y',y" can be pieced together to construct diagonal matrices x and y solving (P_2) for a .

The vector (\mathbf{x}_i) is a fixed point of λ , and its existence is enough to justify the proof of Theorem 2.B.

If a $\in \mathcal{Q}$ is not fully admissible, but is such that (2.3) or (2.9) hold for every proper subset F, then a is said to be <u>rartially admissible</u>. For such a matrix a, there will not exist b $\in \mathcal{Q} \cap \mathfrak{F}(C,R)$ having the same zero pattern as a. (But if we allow b to have extra zeros, then such matrices b will exist.) Thus Theorem 2.D amounts to saying that whenever there exists a matrix in $\mathcal{Q} \cap \mathfrak{F}(C,R)$ having the same zero pattern as a, then x and y exist, and conversely. Hence ϕ defined by (2.1) has critical points in \mathbb{R}^m_+ if and only if a is fully admissible. Finally we note that for the doubly stochastic case, where m=n and $\sigma_1 = \delta_1 = 1$ for each i and j, full admissibility reduces to a cardinality condition.

§3. <u>Invariants</u>

Observe that if π is a k-cycle then

is invariant under C and R , provided it is well-defined. In fact it is invariant under any diagonal scaling xay. These ratios can be used to compute $\lim_{N \to \infty} N$ when it exists, as illustrated by the following example.

Example 3.A:

Let

$$\mathbf{a} = \begin{pmatrix} 0 & \mathbf{a}_{12} & \mathbf{a}_{13} \\ \mathbf{a}_{21} & 0 & \mathbf{a}_{23} \\ \mathbf{a}_{31} & \mathbf{a}_{32} & 0 \end{pmatrix}$$

and $\sigma_{i} = \delta_{j} = 1$ for each i and j . Then $r = a_{12}a_{23}a_{31}a_{13}a_{32}a_{21}^{-1}$

$$r = a_{12}a_{23}a_{31}a_{13}a_{32}a_{21}$$

is invariant. Thus

$$\lim_{N \to \infty} \Lambda^{N}_{\mathbf{a}} = \begin{pmatrix} 0 & \mu & 1-\mu \\ 1-\mu & 0 & \mu \\ \mu & 1-\mu & 0 \end{pmatrix}$$

where $\mu = r^{\frac{1}{3}(1+r^{\frac{1}{3}})^{-1}}$. And a diagonal scaling that leads to this is

$$x = diag(1, \frac{a_{13}}{a_{23}} \frac{\mu}{1-\mu}, \frac{a_{12}}{a_{32}} \frac{1-\mu}{\mu})$$

$$y = diag(\frac{a_{32}}{a_{31}a_{12}}, \frac{\mu^2}{1-\mu}, \frac{\mu}{a_{12}}, \frac{1-\mu}{a_{13}})$$

Finally, for general σ and δ conditions (2.3) become

$$\delta_1 < \sigma_2 + \sigma_3$$
, $\delta_2 < \sigma_1 + \sigma_3$, $\delta_3 < \sigma_1 + \sigma_2$.

From now on we assume that a has a positive row and column. Without loss of generality these may be the first row and column. Then we can define a ratio matrix r by

(3.2)
$$r_{ij} = a_{ij} a_{11} a_{i1}^{-1} a_{ij}^{-1}$$

This matrix is invariant under C and R , and can be used to define analogues of λ

and ϕ . By writing the diagonal matrix x of (P_2) in the form $x_i = t_i a_{i1}^{-1}$ we are led to consider the operator λ defined by

(3.3)
$$(\hat{\lambda}t)_{i} = \left[\sum_{j=1}^{n} r_{ij} \delta_{j} \left(\sum_{k=1}^{m} r_{kj} t_{k}\right)^{-1}\right]^{-1} \sigma_{i}.$$

If $\lambda t=t$ then the matrix b defined by

(3.4)
$$b_{ij} = r_{ij} \left(\sum_{k=1}^{m} r_{kj} t_{k} \right)^{-1} t_{i} \delta_{j} = a_{ij} \left(\sum_{k=1}^{m} a_{kj} t_{k} a_{kl}^{-1} \right) t_{i} a_{il}^{-1} \delta_{j}$$

 $\Im(C,R)$; and, conversely, if b of the form $b_{ij} = a_{ij}t_ia_{i1}^{-1}s_j$ is in $\Im(C,R)$ then And the corresponding relationship between iterates of $\hat{\lambda}$ and Λ is

(3.5)
$$(\Lambda^{N}a)_{ij} = r_{ij} \left[\sum_{k=1}^{m} r_{kj} (\hat{\lambda}^{N-1}\hat{t}^{0})_{k} \right]^{-1} (\hat{\lambda}^{N-1}\hat{t}^{0})_{i} \delta_{j}$$

where

$$\hat{\mathbf{t}}_{\mathbf{i}}^{0} = \mathbf{a}_{\mathbf{i}\mathbf{1}}\mathbf{t}_{\mathbf{i}}^{0} .$$

And by inverting this it follows that

(3.7)
$$(\hat{\lambda}^{N}t^{0})_{i} = a_{i1}o_{i}^{N+1}\prod_{k=0}^{N} [\sum_{j=1}^{n} (\Lambda^{k}a)_{ij}]^{-1}.$$
 Similarly the corresponding function to minimize is

(3.8)
$$\hat{\Phi}(t) = \prod_{j=1}^{n} (\sum_{i=1}^{m} r_{i,j} t_{i})^{\delta_{j}} \prod_{i=1}^{m} t_{i}^{-\delta_{i}}.$$

Problem (P2) amounts to seeking a matrix b with the same ratio matrix as a , having prescribed row and column sums. And the uniqueness result in Theorem 2.A amounts to saying that distinct matrices with identical row and column sums have distinct ratio matrices. Finally, we note that because of (1.1) both ϕ and ϕ are homogeneous in t, and thus it suffices to minimize

(3.9)
$$\Psi(t) = \prod_{j=1}^{n} \left(\sum_{j=1}^{m} a_{ij} t_{i} \right)^{\delta_{j}}$$

or

(3.10)
$$\hat{\Psi}(t) = \prod_{j=1}^{n} \left(\sum_{i=1}^{m} r_{ij} t_{i} \right)^{\delta_{j}}$$

over R, subject to

The solution to the former is

(3.12)
$$\mathbf{t}_{i}^{*} = \mathbf{c} \prod_{k=0}^{\infty} \sigma_{i} \left[\sum_{j=1}^{n} (\Lambda^{k} \mathbf{a})_{ij} \right]^{-1}$$

and the solution to the latter is

(3.13)
$$\hat{t}_{i}^{*} = \hat{c} a_{i1} t_{i}^{*}$$

where c and c are appropriate scale factors. The remainder of this section is devoted to examples.

Example 3.B:

For the case m=n=2 the limit is

$$\begin{pmatrix}
\mu & \sigma_1^{-\mu} \\
\delta_1^{-\mu} & \mu + \sigma_2^{-\delta_1}
\end{pmatrix}$$

where μ is a nonnegative solution of

$$\mu^{2} - (\delta_{1} + \frac{r_{22}\sigma_{1} + \sigma_{2}}{r_{22}^{-1}})\mu + \frac{r_{22}\sigma_{1}\delta_{1}}{r_{22}^{-1}} = 0$$

provided a is nonsingular. Otherwise $\mu = \sigma_1 \delta_1 (\sigma_1 + \sigma_2)^{-1}$. If $a_{22} = 0$ the condition (2.3) becomes $\delta_2 < \sigma_1$, and if $\delta_2 = \sigma_1$ then $\mu = 0$. The corresponding variational problem is to

minimize
$$(s+1)^{\delta_1}(s+r_{22})^{\delta_2}s^{-\sigma_1}$$
.

 $s = t_1 \frac{\sigma_1 + \sigma_2}{\sigma_2}$

Example 3.C:

For the case m=2, n=3 the limit is

$$\begin{pmatrix}
\mu & \eta & \sigma_1 - \mu - \eta \\
\delta_1 - \mu & \delta_2 - \eta & \mu + \eta + \delta_3 - \sigma_1
\end{pmatrix}$$

where μ, η are nonnegative solutions of

$$\mu \eta + \frac{\delta_2}{r_{22}^{-1}} \mu - \frac{r_{22}^{\delta_1}}{r_{22}^{-1}} \eta = 0$$

$$\mu^{2} + \mu \eta - (\sigma_{1}^{2} + \frac{r_{32}^{\delta_{1} + \delta_{3}}}{r_{32}^{-1}})\mu - \frac{r_{32}^{\delta_{1}}}{r_{32}^{-1}}\eta + \frac{r_{32}^{\sigma_{1}^{\delta_{1}}}}{r_{32}^{-1}} = 0.$$

These reduce to a single cubic in μ . If $r_{22}=1$ then the first equation becomes $\eta=\delta_2\delta_1^{-1}\mu$. If $r_{32}=1$ then the second equation reduces to $(\delta_1+\delta_3)\mu+\delta_3\eta=\sigma_1\delta_1$. And if $r_{22}=r_{23}=1$ then $\mu=\sigma_1\delta_1(\sigma_1+\sigma_2)^{-1}$ and $\eta=\sigma_1\delta_2(\sigma_1+\sigma_2)^{-1}$. The corresponding variational problem is to

minimize
$$(s+1)^{\delta_1}(s+r_{22})^{\delta_2}(s+r_{23})^{\delta_3}s^{-\sigma_1}$$

Here, as before,
$$s = t_1^{\frac{\sigma_1 + \sigma_2}{\sigma_2}}$$

Example 3.D:

For the case m=n=3, $\sigma_i = \delta_i = 1$ for each i and j; if $r_{23} = r_{32} = r_{22} r_{33}$ or if $r_{22}^{=r_{33}^{=r_{23}^{r_{32}}}}$ then the limit is a permuted matrix (i.e. all rows and columns are permutations of one another). In the former case it is symmetric; in the latter, skewsymmetric.

Example 3.E:

Let

$$\mathbf{a} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \end{pmatrix}$$

The conditions (2.3) become

$$\sigma_2 < \delta_1 + \delta_4$$
, $\sigma_4 < \delta_1 + \delta_2$, $\delta_2 < \sigma_1 + \sigma_3$

 $\sigma_2 < \delta_1 + \delta_4 \text{ , } \sigma_4 < \delta_1 + \delta_2 \text{ , } \delta_2 < \sigma_1 + \sigma_3 \text{ .}$ Thus we can consider the case $\sigma_i = \delta_j = 1$ for each i and j . The corresponding variational problem is to

minimize
$$(t_1+t_2+t_3+t_4)(t_1+t_3)(t_1+t_2+t_3)(t_1+t_3+t_4)$$

subject to $t_1t_2t_3t_4=1$.

By symmetry $t_1 = t_3$ and $t_2 = t_4$. Thus we need to minimize

$$(t_1^2 + 1)(2t_1 + t_1^{-1})^2$$
.

Letting $s = t_1^2$ the condition for a minimum is $8s^3 + 8s^2 = 1$. The only positive solution is $\frac{1}{4}(\sqrt{5}-1)$. Thus

$$t_1 = t_3 = \frac{1}{2}\sqrt{5}-1$$
 , $t_2 = t_4 = \sqrt{5}+1$.

This means that diagonal scaling matrices are

$$x = diag(t_1, t_2, t_3, t_4)$$

 $y = \frac{1}{2} diag(\sqrt{5\sqrt{5}-11}, t_2, \sqrt{2\sqrt{5}-4}, \sqrt{2\sqrt{5}-4})$

and the limit xay is

$$\begin{pmatrix}
g - \frac{1}{2} & \frac{1}{2} & \frac{1}{2} - \frac{1}{2}g & \frac{1}{2} - \frac{1}{2}g \\
1 - g & 0 & 0 & g \\
g - \frac{1}{2} & \frac{1}{2} & \frac{1}{2} - \frac{1}{2}g & \frac{1}{2} - \frac{1}{2}g \\
1 - g & 0 & g & 0
\end{pmatrix}$$

where $g = \frac{1}{2} (\sqrt{5} - 1)$ is the golden mean.

§4. Scaling Infinite Matrices

In this section we consider the scaling of infinite matrices; i.e. one or both of m and n is infinite. We assume $n=\infty$, and that $\sigma\in\mathbb{R}^m$ and $\delta\in\ell_1$, are positive vectors satisfying (1.1). Denote by ℓ_p^m the Banach space of infinite $m\times\infty$ -matrices a having real entries such that

Note that ℓ_2^m is a Hilbert space for any m (even $m=\infty$). Let $a \in \ell_1^m$ be positive, and consider the following analogue of (P_2) :

Find $t \in \mathbb{R}_{+}^{m}$ and $s \in \mathbb{R}_{+}^{\infty}$ such that the matrix b given by $b_{ij} = a_{ij}t_{i}s_{j}$

$$\sum_{j=1}^{\infty} b_{ij} = \sigma_i$$
, $i = 1,...,m$; $\sum_{i=1}^{m} b_{ij} = \delta_j$, $j = 1,2,...$.

The matrix $b \in \ell_1^m$ is said to be a solution of (P_2) .

Theorem 4.A

(P2)' has at most one solution.

Proof:

As in the proof of Theorem 2.A, we assume that \hat{a} and $(\hat{t}_i\hat{a}_{ij}\hat{s}_j)$ are solutions of (P)', and show that $\hat{t}_i = t_1$; i = 1, ..., m. For each i = 1, ..., m

(4.2)
$$\hat{t}_{i} = \sum_{k=1}^{m} \alpha_{ik} \hat{t}_{k}$$

where $\alpha \in \mathbb{R}_{+}^{m,m}$ is given by

(4.3)
$$\alpha_{ik} = \sigma_i^{-1} \hat{\tau}_i \sum_{j=1}^{\infty} \hat{a}_{ij} \hat{a}_{kj} \delta_j^{-1} \hat{s}_j.$$

We note that α satisfies

(4.4)
$$\sum_{k=1}^{m} \alpha_{ik} = 1 , i = 1,...,m ; \sum_{i=1}^{m} \alpha_{ik} \sigma_{i} = \sigma_{k} , k = 1,...,m .$$

By Jensen's inequality

(4.5)
$$\gamma_{i} = e^{-t_{i}} \leq \sum_{k=1}^{m} \alpha_{ik} \gamma_{k} ; \quad i = 1, ..., m.$$

If $m = \infty$, note that the sum on the left of (4.5) converges, as $\gamma_i \le 1$. We have

(4.6)
$$\sum_{i=1}^{m} \sigma_{i} \gamma_{i} \leq \sum_{k=1}^{m} \sum_{i=1}^{m} \alpha_{ik} \sigma_{i} \gamma_{k} = \sum_{k=1}^{m} \sigma_{k} \gamma_{k},$$

and so

(4.7)
$$\gamma_{i} = \sum_{k=1}^{m} \alpha_{ik} \gamma_{k}$$
; $i = 1,...,m$.

But of course the same argument can be used to show that

(4.8)
$$\gamma_{i}^{n} = \sum_{k=1}^{m} \alpha_{ik} \gamma_{k}^{n}$$
; $i = 1,...,m$

for any integer $n \ge 0$. And by taking limits of sums it follows that

(4.9)
$$e^{-(\gamma_{i}-\gamma_{1})^{2}} = \sum_{k=1}^{m} \alpha_{ik} e^{-(\gamma_{k}-\gamma_{1})^{2}}.$$

If
$$t_i \neq t_1$$
 for some i , then for all $i = 1, ..., m$

$$\sum_{k=1}^{m} \frac{-(\gamma_k - \gamma_1)^2}{\alpha_i k^e} < 1.$$

Since, by (4.9), the inequality (4.10) cannot hold for i=1, it must indeed be true that t is a constant vector.

To consider the question of existence, suppose first that $m < \infty$. For $N \ge 1$ let $\rho_N = \sum_{i=1}^m \sigma_i / \sum_{j=1}^N \delta_j$. According to Theorem 2.A there is a unique matrix $a^N \in \mathbb{R}_+^{m,N}$ such that $a_{ij}^{N} = a_{ij}^{N} t_{i}^{N} s_{i}^{N}$ and

Define

(4.12)
$$b_{ij}^{N} = \begin{cases} a_{ij}^{N}, & j \leq N \\ 0, & j \geq N \end{cases}$$

Theorem 4.B:

For the case $m < \infty$, the sequence $\{b^N\}$ defined by (4.12) converges in the norm topology of ℓ_1^m to the solution b of (P_2) .

Proof:

The sequence $\{b^N\}$ is bounded in ℓ_2^m and as such has a subsequence $\{b^N\}$ which converges weakly to an element $b \in \ell_2^m$. Clearly $b_{ij} \geq 0$; $i = 1, \ldots, m$, $j = 1, 2, \ldots$. Define functionals f_j on ℓ_2^m by

(4.13)
$$f_{j}(z) = \sum_{i=1}^{m} z_{ij}, z \in \ell_{2}^{m}$$

By taking weak limits it follows that $f_j(b) = \delta_j$. Thus $b \in \ell_1^m$ and $\{b^k\}$ converges to b in the norm topology of ℓ_1^m . And $\sum_{j=1}^{\infty} b_{ij} = \sigma_i$; $i = 1, \ldots, m$. As in §3 define r_{ij} by (3.2). Then, again by taking limits,

(4.14)
$$b_{ij}b_{11} = r_{ij}b_{i1}b_{ij}$$
; $i = 1,...,m, j = 1,2,...$

And this shows that b is a solution to (P2)'.

If we write $b_{ij} = a_{ij}t_{i}s_{j}$ then, as before, $\lambda t = t$, where λ is given by (1.5). And if R,C and Λ are defined on ℓ_1^m by (1.2) and (1.3), then the same proof used for Theorem 2.B can be used here to establish the following result.

Theorem 4.C:

The sequence $\{\Lambda^N a\}$ converges in ℓ_1^m to the solution, b, of (P_2) .

For the case $m=\infty$ the solution to $(P_2)^i$ can be constructed as follows. For $n\geq 1$ set $\kappa_N = \sum\limits_{j=1}^\infty \delta_j / \sum\limits_{i=1}^N \sigma_i$. According to Theorem 4.A there is a unique $a^N \in \ell_1^N$ of the form $a_{i,j}^N = a_{i,j} \ell_i^N s_j^N$ such that

As before, define

(4.16)
$$b_{ij}^{N} = \begin{cases} a_{ij}^{N} & , & i \leq N \\ & & \\ & & \\ & & \\ & & & \end{cases}$$

An argument similar to the one used to prove Theorem 4.B establishes the following result.

Theorem 4.D:

For the case $m = \infty$, the sequence $\{b^N\}$ defined by (4.16) converges in the norm topology of ℓ_1^∞ to the solution b of (P_2) .

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Certain invariants under diagonal scaling are used to actually compute the desired scaled matrix, and examples are provided. Finally, at the end of the paper, a

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